# NH winter forecast skill of AO and NAO indices: results and sampling issues

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Reusing material from earlier talks given together with

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## **Outline**

Intro: ECMWF System 4

Predicting NH winter circulation modes

Challenges of sampling

Discussion



# System 4 configuration

IFS: T<sub>L</sub>255L91 Cy36r4

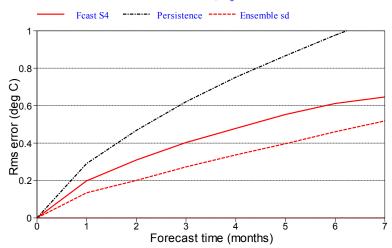
- Real time forecasts:
  - 51 member ensemble forecast to 7 months
  - SST and atmos. perturbations added to each member
- Back integrations from 1981-2010 (30 years)
  - 15 member ensemble every month
  - 15 members extended to 13 months once per quarter
  - 51 members for Feb/May/Aug/Nov starts



# ENSO forecasts are good .....

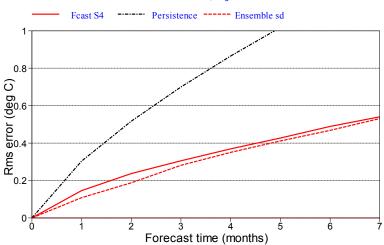
#### NINO3.4 SST rms errors

95% confidence interval for 0001, for given set of start dates



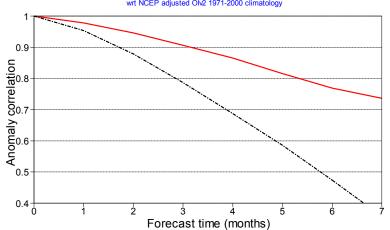
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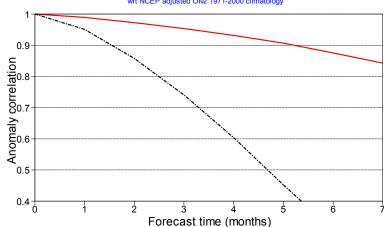


1981-1995

NINO3.4 SST anomaly correlation



#### NINO3.4 SST anomaly correlation



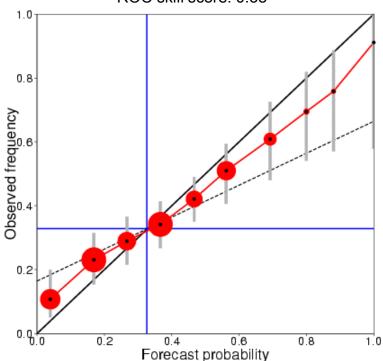


1996-2010

# So are probabilistic scores ....

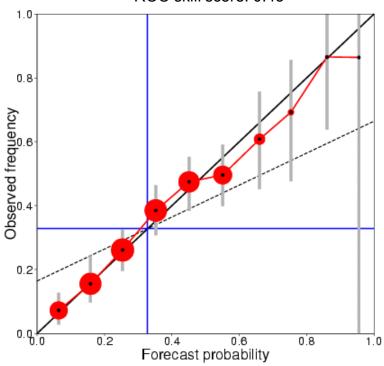
#### 15 members

JJA Europe T2m>upper tercile
Re-forecasts from 1 May, 1981-2010
Reliability score: 0.987
ROC skill score: 0.38



#### 51 members

JJA Europe T2m>upper tercile Re-forecasts from 1 May, 1981-2010 Reliability score: 0.996 ROC skill score: 0.43



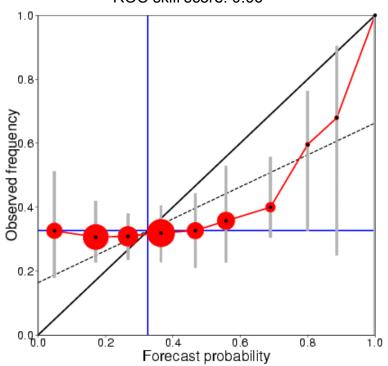
(Figures from Susanna Corti)



## **Ensemble size important for low-signal areas**

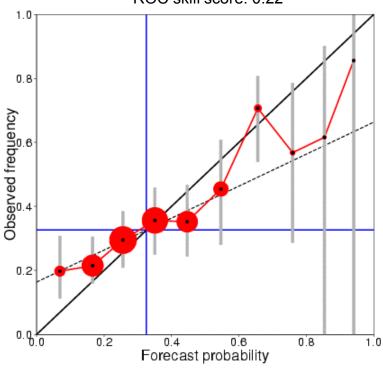
#### 15 members

DJF Europe T2m>upper tercile Re-forecasts from 1 Nov, 1981-2010 Reliability score: 0.902 ROC skill score: 0.06



#### 51 members

DJF Europe T2m>upper tercile Re-forecasts from 1 Nov, 1981-2010 Reliability score: 0.981 ROC skill score: 0.22



(Figures from Susanna Corti)



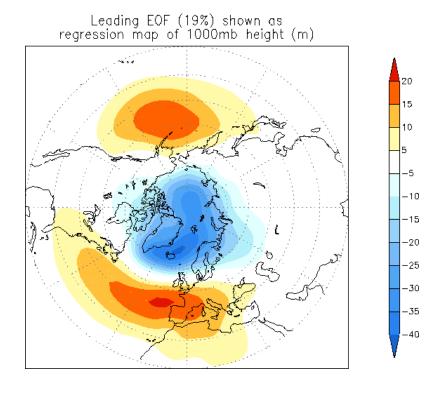
## **Arctic Oscillation**

Calculated as first EOF of monthly mean MSLP anomalies, poleward of 20N.

Use same method as CPC, but using ERA interim analysis, 1981-2010.

Model and analysis time-series both obtained by projection onto **observed** EOF.

Correlation of our observed time-series with CPC is 0.996.



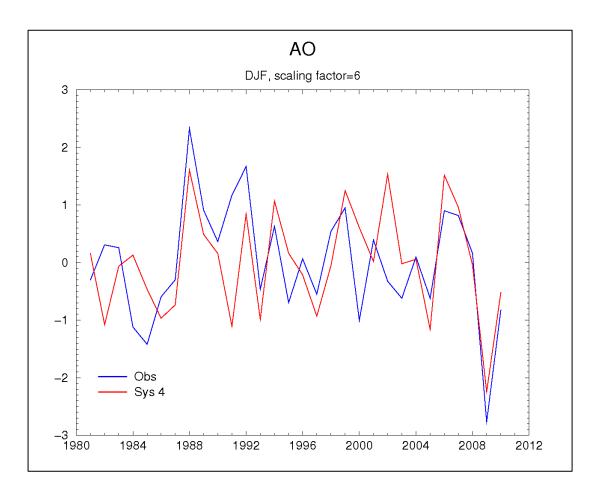
**EOF** (from CPC)



## **AO** re-forecast skill

**Correlation (30y) = 0.608** 

26 years (no volcanoes) Correlation = 0.73



Surprising because model AO is very noisy ....



# Statistical analysis

```
Unbiased variance estimates: Obs/Tot/Int/Ext:
                                             1.0000
                                                        0.8390
                                                                  0.8316
                                                                             0.0074
Model/obs stddev ratio:
Model/obs stddev ratio interval:
                                                     ← model variability consistent with obs
                              0.693 1.129
Bootstrap over nens, pval for ratio=1: 0.7960
                               0.0941
SNR actual
SNR jackknife over nens
                               0.0202 0.1029 0.1857
______
                               0.6085
ACC actual
ACC basic bootstrap over nens : 0.5568 0.7121 0.8144
                                                     ← 95% interval due to ensemble size
                               ACC basic bootstrap over nyears:
ACP from internal sampling: -0.2947 0.0583 0.4010
Mean ACC for nens-1:
                  0.6049
                                              ← only a 0.0004 chance we could get this correlation
p val of measured acc if model perfect:
                                     0.9996
```

- Model skill for these years is relatively high
- Model predictability limit must be wrong (because we exceed it so much)



# Other teleconnection patterns

	ACC	S/N	ACP	P-val
PNA (EOF)	0.696	0.64	0.54	0.065
NAO (EOF)	0.465	0.13	0.10	0.017

**PNA** has high skill and high predictability **NAO** has moderate skill, and low predictability

NAO skill is, like AO, higher than expected



# **Does resolution help?**

**Project Minerva** has run the ECMWF coupled model at different atmospheric resolutions. We have 30 years of winter forecasts, with 51 member ensembles:

	T319		T639	
	ACC	S/N	ACC	S/N
PNA (EOF)	0.68	0.69	0.69	0.73
NAO (EOF)	0.36	0.17	0.63	0.18

**S/N** does not seem to be affected by resolution.

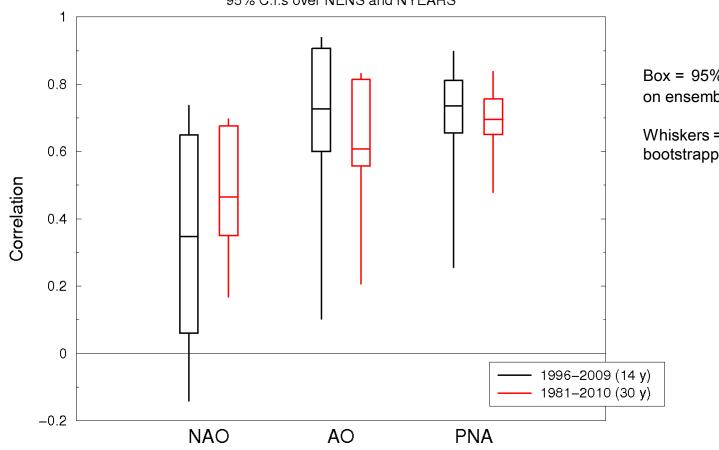
**NAO** structure and skill is significantly (at 5% level) improved by higher atmosphere resolution.



# Challenge: sampling errors are large!

#### Correlation scores, ECMWF S4



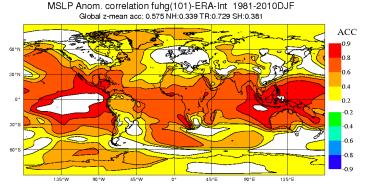


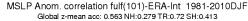
Box = 95% interval, bootstrapping on ensemble size

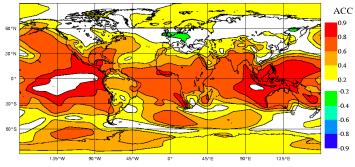
Whiskers = 95% interval, bootstrapping on years included



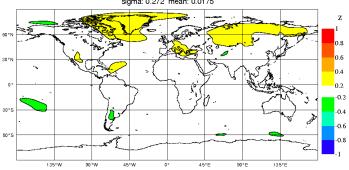
## **NH** winter forecasts



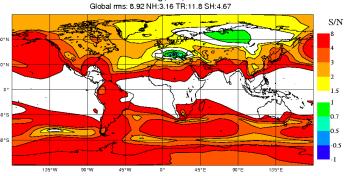




Fisher z transform diff fuhg(101)-fulf(101) 1981-2010DJF sigma: 0.272 mean: 0.0175



MSLP Ens. mean S/N ratio fuhg(101)-ERA-Int 1981-2010DJF Global rms: 8.92 NH:3.16 TR:11.8 SH:4.67



Even with 101 members, ensemble mean signal not always well defined



## **Conclusions**

## S4 has substantial skill in predicting AO phase over a 30 year period

- How typical this is of expected future performance is unknown
- Amplitude of model signal is too weak
- Models are noisy

#### Scores are unstable

- Sensitive to choice of years, especially for shorter periods
- Relative skill of AO and NAO indices can vary between model versions

# Higher resolution (to T639)

- DOES help NAO in particular (quite big improvement)
- Does **NOT** help S/N ratios
- ... according to a single experiment



# **Conclusions - Sampling**

## Sampling over NYEARS

- Is an obvious problem for systems without high S/N ratios
- Skill estimates need as many years as possible, but there are limits
- We need to understand sources of skill to know how far back we can go (to 1979? to 1960? Even earlier??)

### Ensemble size is often too small

 Given how noisy our models are, we should probably be doing our experiments with ensembles O(100) to get clean results

#### Costs

 So all we need are very high resolution models, large ensembles, lots of start dates ... and lots of different experiments to improve our models.

